RANDOM FOREST REGRESSION

```
In [1]:
         import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         import seaborn as sns
In [2]:
         data = pd.read csv('pima.csv')
         data.head()
           Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age Outcome
         0
                    6
                          148
                                        72
                                                     35
                                                             0 33.6
                                                                                      0.627
                                                                                             50
                           85
                                                     29
                                                             0 26.6
                                                                                      0.351
                                                                                             31
                                                                                                       0
         2
                          183
                                        64
                                                      0
                                                             0 23.3
                                                                                      0.672
                                                                                             32
                    8
         3
                           89
                                         66
                                                     23
                                                             94 28.1
                                                                                             21
                                                                                                       0
                                                                                      0.167
                    0
         4
                          137
                                         40
                                                     35
                                                            168 43.1
                                                                                      2.288
                                                                                            33
In [5]:
         from sklearn.model_selection import train_test_split
         from sklearn.ensemble import RandomForestClassifier
         from sklearn.datasets import make_classification
         from sklearn.metrics import accuracy_score
         from sklearn.preprocessing import StandardScaler, MinMaxScaler
         import pandas profiling
         from matplotlib import rcParams
         import warnings
In [4]:
         pip install pandas_profiling
        Defaulting to user installation because normal site-packages is not writeable
        Collecting pandas_profiling
          Downloading pandas_profiling-3.6.1-py2.py3-none-any.whl (328 kB)
                                                    -- 328.5/328.5 kB 2.3 MB/s eta 0:00:00m eta 0:00:01[36m0:00:01
         Requirement already satisfied: numpy<1.24,>=1.16.0 in /usr/lib/python3/dist-packages (from pandas_profiling) (1.21.5)
         Collecting htmlmin==0.1.12
          Downloading htmlmin-0.1.12.tar.gz (19 kB)
          Preparing metadata (setup.pv) ... done
```

Note: you may need to restart the kernel to use updated packages.

Glucose 0.252459 BMI 0.167545

```
In [6]: data.columns
In [7]:
         X=data.drop("Outcome",axis=1)
         y=data["Outcome"]
 In [8]:
         scaler=StandardScaler()
         X_scaled=scaler.fit_transform(X)
          X\_train, X\_test, Y\_train, Y\_test=train\_test\_split (X\_scaled, y, stratify=y, test\_size=0.10, random\_state=34)
         classifier = RandomForestClassifier(n_estimators=100)
         classifier.fit(X_train,Y_train)
 Out[9]: RandomForestClassifier()
        In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
        On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.
In [10]:
         y_pred = classifier.predict(X_test)
         print("Accuracy:",accuracy_score(Y_test,y_pred))
         feature_importances_df = pd.DataFrame(
             {"feature":list(X.columns),"importance":classifier.feature_importances_}
         ).sort_values("importance",ascending=False)
         feature_importances_df
         Accuracy: 0.8311688311688312
Out[10]:
                        feature importance
```

```
/.sort_values( importance ,ascending=raise)
           feature_importances_df
          Accuracy: 0.8311688311688312
Out[10]:
                           feature importance
          1
                           Glucose
                                     0.252459
          5
                              BMI
                                     0.167545
          7
                              Age
                                     0.136989
          6 DiabetesPedigreeFunction
                                     0.129871
          2
                      BloodPressure
                                     0.088367
          0
                        Pregnancies
                                     0.079612
          4
                            Insulin
                                     0.073444
          3
                      SkinThickness
                                     0.071712
In [11]:
           from sklearn.tree import DecisionTreeClassifier
           clf=DecisionTreeClassifier()
           clf.fit(X_train,Y_train)
           Y_pred = clf.predict(X_test)
           from sklearn.metrics import accuracy_score
           print("Accuracy-DecisionTree :",accuracy_score(Y_test,Y_pred))
          Accuracy-DecisionTree : 0.6753246753246753
```

In []:

NAVIE BAYES

```
In [1]:
          import pandas as pd
          import numpy as np
In [3]:
          data = pd.read_csv('covid_ds.csv')
          data
                            wbc
                                                              ldh
                                                                   diagnosis
Out[3]:
             no
                                      mc
                                              ast
                                                       bc
                      рс
               1
                    Low
                             Low
                                     Low
                                             High Normal Normal
                                                                        True
                     Low
                             Low
                                  Normal
                                             High
                                                   Normal
                                                             High
                                                                        True
          2
               3
                     Low
                            High
                                  Normal
                                             High Normal
                                                          Normal
                                                                       False
          3
               4
                     Low
                            High
                                  Normal
                                             High
                                                     High
                                                          Normal
                                                                        True
           4
               5
                    Low
                          Normal
                                    High
                                             High Normal
                                                          Normal
                                                                       False
          5
               6
                    Low
                                  Normal
                                             High
                                                   Normal
                                                             High
                                                                        True
                          Normal
           6
               7
                  Normal
                             Low
                                     Low
                                             High
                                                   Normal
                                                          Normal
                                                                        True
               8
                  Normal
                            High
                                  Normal
                                             High Normal Normal
                                                                        False
           8
               9
                  Normal
                            High
                                  Normal
                                             High
                                                     High
                                                             High
                                                                        True
          9
              10
                  Normal
                          Normal
                                    High
                                             High Normal Normal
                                                                        False
          10
              11
                  Normal
                          Normal
                                    High
                                             High Normal
                                                             High
                                                                        True
              12
                    High
                                                                        True
          11
                             Low
                                     Low
                                          Normal
                                                  Normal Normal
          12
              13
                    High
                          Normal
                                    High
                                          Normal
                                                  Normal
                                                          Normal
                                                                       False
          13
                    High
                          Normal
                                    High
                                          Normal
                                                     High
                                                             High
                                                                        True
          14
              15
                    High
                            High
                                  Normal
                                          Normal Normal
                                                             High
                                                                        True
          15
              16
                     Low
                          Normal
                                    High
                                             High
                                                     High Normal
                                                                        False
              17
                 Normal
                                                     High Normal
                                                                       False
          16
                          Normal
                                    High
                                            High
                           Low
                                    Low Normal Normal
```

```
Low Normal
        15 16
                                High
                                        High
                                               High Normal
                                                                False
        16 17 Normal Normal
                                        High
                                                                False
                                High
                                               High Normal
        17 18
                  High
                                 Low Normal Normal
                                                                True
                                        High Normal Normal
        18 19 Normal Normal Normal
                                                                False
        19 20 Normal
                         High Normal
                                        High Normal
                                                      High
                                                                True
        20 21 Normal
                          Low Normal
                                        High Normal
                                                      High
                                                                True
        21 22
                  Low
                         High Normal
                                        High
                                               High
                                                      High
                                                                True
        22 23
                                        High
                                               High
                                                      High
                                                                True
                  Low
                          Low
                                 Low
        23 24
                  High
                         High Normal Normal Normal
                                                                True
        24 25
                  High Normal Normal Normal Normal
                                                                False
In [4]:
         from sklearn import preprocessing
         le = preprocessing.LabelEncoder()
         pc_encoded=le.fit_transform(data['pc'].values)
         wbc_encoded=le.fit_transform(data['wbc'].values)
         mc_encoded=le.fit_transform(data['mc'].values)
         ast_encoded=le.fit_transform(data['ast'].values)
         bc_encoded=le.fit_transform(data['bc'].values)
         ldh_encoded=le.fit_transform(data['ldh'].values)
         Y=le.fit_transform(data['diagnosis'].values)
In [5]:
         X=np.array(list(zip(pc_encoded,wbc_encoded,mc_encoded,ast_encoded,bc_encoded,ldh_encoded)))
         Χ
         Υ
Out[5]: array([1, 1, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1,
               1, 1, 0])
In [6]:
         from sklearn.naive_bayes import MultinomialNB
         from sklearn.metrics import accuracy_score
         from sklearn.metrics import classification_report
         model = MultinomialNB()
```

High

True

14 15

High

High Normal Normal Normal

```
In [7]:
         from sklearn.model_selection import train_test_split
         X_train,X_test,Y_train,Y_test=train_test_split(X,Y)
In [8]:
         model.fit(X_train, Y_train)
         y_pred = model.predict(X_test)
         print("Accuracy:",accuracy_score(Y_test, y_pred))
         print("\nReport")
         print(classification_report(Y_test,y_pred))
        Accuracy: 0.7142857142857143
        Report
                      precision
                                  recall f1-score
                                                   support
                                    0.33
                                                           3
                   0
                          1.00
                                              0.50
                                    1.00
                   1
                          0.67
                                              0.80
                                                           4
                                                          7
                                              0.71
            accuracy
           macro avg
                          0.83
                                    0.67
                                              0.65
                                                          7
        weighted avg
                          0.81
                                    0.71
                                              0.67
                                                           7
```

SVM

```
In [2]:
         from sklearn.svm import SVC
         from sklearn import svm
         import numpy as np
         X=np.array([[3,4],[1,4],[2,3],[6,-1],[7,-1],[5,-3]])
         y=np.array([-1,-1,-1,1,1,1])
         l=SVC(C=1e5,kernel='linear')
         1.fit(X,y)
         print('w= ',l.coef )
         print('b= ',1.intercept_)
         print('Indices of support vectors= ',l.support_)
         print('Support vectors= ',1.support_vectors_)
         print('No. of support vectors from each class= ',l.n support )
         print('coefficient of support vectors in decision function= ',np.abs(l.dual_coef_))
         import pandas as pd
         data=pd.read csv('glass.csv')
         data.head()
         x=data.drop('Type',axis=1)
         y=data.Type
         from sklearn.model selection import train test split
         x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
         linear=svm.SVC(kernel='linear')
         linear.fit(x_train,y_train)
         print(linear.support vectors )
         print(linear.n_support_)
         y_pred=linear.predict(x_test)
         from sklearn.metrics import accuracy score
         print(accuracy_score(y_test,y_pred))
         from sklearn.metrics import confusion matrix
         print(confusion_matrix(y_test,y_pred))
         from sklearn.metrics import classification_report
         print(classification_report(y_test,y_pred))
```

```
model1=SVC(kernel='sigmoid')
model2=SVC(kernel='poly')
model3=SVC(kernel='rbf')
model1.fit(x train,y train)
model2.fit(x_train,y_train)
model3.fit(x_train,y_train)
y pred1=model1.predict(x test)
y pred2=model2.predict(x test)
y pred3=model3.predict(x test)
print("prediction by model1 ",accuracy_score(y_test,y_pred1))
print("prediction by model2",accuracy score(y test,y pred2))
print("prediction by model3",accuracy_score(y_test,y_pred1))
W = [[0.25 - 0.25]]
b = [-0.75]
Indices of support vectors= [2 3]
Support vectors= [[ 2. 3.]
[ 6. -1.]]
No. of support vectors from each class= [1 1]
```

coefficient of cumport vectors in decision function- [[A AG25 A AG25]]